Short-term Planning, Monetary Policy, and Macroeconomic Persistence

Christopher Gust Edward Herbst David López-Salido* (Federal Reserve Board)

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^{*}The views expressed are solely our responsibility and should not be interpreted as reflecting the views of the Board of Governors of the Federal Reserve System or of anyone else associated with the Federal Reserve System.

Motivation: Behavioral Macro Models

- Recent behavioral macro models emphasize that agents' expectations can be rooted in human judgement and experimental evidence instead of being assumed fully-rational.
 - ► Gabaix's (2018) limited attention model.
 - ► Angeletos and Lian's (2016) lack of common knowledge.
 - Farhi and Werning's (2017) k-level thinking.
 - ► Woodford's (2018) finite planning horizons (FH).

▶ Do "new behavioral" models provide empirically-realistic macro dynamic to study the effects of monetary policy?

Motivation: Macroeconomic Persistence

"The pervasiveness of sluggish responses in the macro data, combined with the implausibility of many of the micro stories underlying adjustment cost models, suggests that we look for a different approach to modeling the sources of inertia in both prices and real variables."

Sims (1998), Stickiness.

Plan of the Presentation

- Heuristic description of finite-horizon planning.
- Preview of the results.
- ▶ Formal representation of the aggregate equilibrium.
 - Microeconomic heterogeneity.
 - Value function updating and trend-cycle decomposition.
 - Short-term planning and monetary policy.
 - A new trend-cycle decomposition.
 - A (modified) Taylor principle.
- ► Estimation results.
 - Two key parameters.
 - Trend-cycle decomposition of US output, inflation, and short-term rate.
 - ▶ Individual heterogeneity: disagreement of expectations.

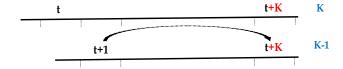
Finite-Horizon Planning: Heuristic

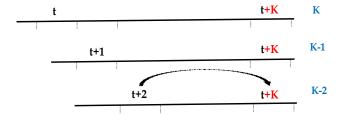
Finite-Horizon (FH) Planning

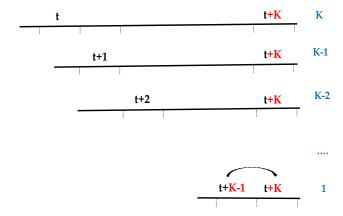
- ▶ The backbone of the model is NK.
- Agents make plans over a finite horizon (FH):
 - ► Too costly to search all possible decision tree (infinite-horizon state-contingent plan, "Borges' garden of forking paths")
 - They transform an infinite-horizon problem into a sequence of shorter finite-horizon ones (finite future.)
 - They are "boundedly rational" in thinking about the continuation values of their plans.
 - ► They (learn) update their beliefs on the continuation values of their plans based on past data/experience.
- Agents are forward-looking in thinking about events over their planning horizon, but are also backward looking in thinking about events beyond that point.
- ► HANK: Rich cross-sectional heterogeneity in the length of planning horizons.

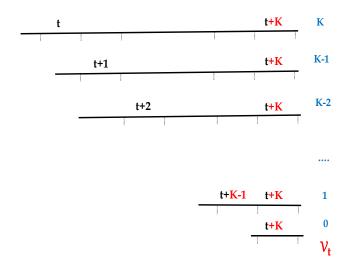


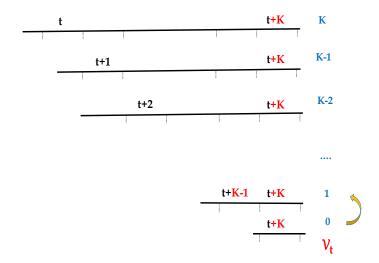


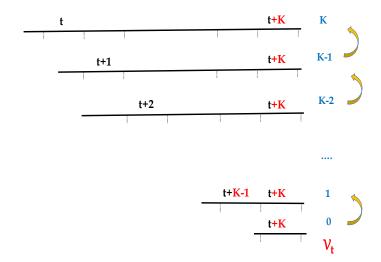












Macroeconomic Persistence

The **model generates persistence** through a novel "trend-cycle decomposition."

- ► The "cyclical component" depends on agents' forward looking behavior over their planning period (i.e., absent learning).
- ► The "trend component" reflects how agents update beliefs about their continuation plans (i.e., value functions) to past information.
 - Without habits in consumption, indexation clauses, and interest-rate smoothing.
 - Without purely "backward-looking expectations."

Preview of the Results

Preview of the Results (1/2)

- We estimate the FH model using U.S. quarterly data on output, inflation, and interest rates from 1966 until 2007.
 - About 50 percent of agents plans for the current-quarter, and a small fraction have planning horizons beyond 2yrs.
 - Agents update their value functions slowly in response to incoming data.
- ► Model goodness of fit is substantially better than:
 - ► The hybrid-NK model and other behavioral macro models (such as *Angeletos and Lian's (2016)* and *Gabaix's (2018)*.)

Preview of the Results (2/2)

- Model generates "substantial persistence" in output, inflation, and interest rates.
 - Without any of the usual mechanisms and without "interest-rate smoothing."
- Business cycle matches "conventional wisdom."
- Measure of trend inflation displays similar movements to the SPF measure of longer-term inflation expectations.
- "Disagreement about inflation expectations" due to agents' heterogeneous plans matches the contour derived from the Michigan Survey.

The Formal Model

Households K-Horizon Planning

► HH problem in t is to choose state-contingent plan $C_{\tau}(s_{\tau})$ for periods $t \leq \tau \leq t + k$ to maximize:

$$\mathbb{E}_{t}^{k} \left[\sum_{\tau=t}^{t+k} \beta^{\tau-t} u(C_{\tau}, \xi_{\tau}) + \beta^{k+1} \vartheta(B_{t+k+1}, s_{t+k}) \right]$$

subject to the budget constraint

$$B_{\tau+1} = (1+i_{\tau})[B_{\tau+1}/\Pi_{\tau} + Y_{\tau} - C_{\tau}]$$

- ▶ The nominal value of government debt maturing in period τ is deflated by the aggregate price level $P_{\tau-1}$.
- ▶ $\vartheta(B_{\tau+1}, s_{\tau})$ is the value function used by the HH to evaluate situations in final state τ .
- ▶ $\mathbb{E}_{t}^{k}[\cdot]$ is the expectation at time t for agents with k forward planning horizon.

Households K-Horizon Planning

▶ **Optimal plan** for $t \le \tau \le t + k$

$$\begin{array}{lcl} u_c(\mathit{C}_{\tau}, \xi_{\tau}) & = & \beta \mathbb{E}_t^{\mathit{k}} \left[\frac{(1+\mathit{i}_{\tau})}{\Pi_{\tau+1}} u_c(\mathit{C}_{\tau+1}, \xi_{\tau+1}) \big| \mathit{s}_{\tau} \right] \text{ and} \\ u_c(\mathit{C}_{t+\mathit{k}}, \xi_{t+\mathit{k}}) & = & \beta (1+\mathit{i}_{t+\mathit{k}}) \vartheta_B(\mathit{B}_{t+\mathit{k}+1}, \mathit{s}_{t+\mathit{k}}) \end{array}$$

- **Expectations** in $t \le \tau \le t + k$ used in planning:
 - Understand the model structure during the planning.
 - For any j periods between t and t+k, aggregate conditions in t+j assumed to be determined by k-j horizon forward looking HHs

$$t \leq \tau \leq t + k: \qquad \mathbb{E}_{t}^{k} \{ Z_{\tau} | s_{\tau} \} = E_{t} \{ Z_{\tau}^{t+k-\tau} \}$$

$$t+1 \leq \tau \leq t + k: \qquad \mathbb{E}_{t}^{k} \{ Z_{\tau+1} | s_{\tau} \} = E_{\tau} \{ Z_{\tau+1}^{t+k-\tau} \}$$

Assume the same planning horizon for all others.

Households K-Horizon Planning

▶ Optimal plan for $t \le \tau \le t + k$ and $1 \le j \le k$ the intertermporal decision is given by:

$$u_{c}(C_{\tau}^{j}, \xi_{\tau}) = \beta E_{\tau} \left[\frac{(1+i_{\tau}^{j})}{\prod_{\tau+1}^{j-1}} u_{c}(C_{\tau+1}^{j-1}, \xi_{\tau+1}) \right]$$

$$u_{c}(C_{\tau}^{0}, \xi_{\tau}) = \beta (1+i_{\tau}) \vartheta_{B}(B_{\tau+1}^{0}; s_{\tau})$$

▶ **Log-linear approx.** (constant $\vartheta(B) = \frac{u(C,\xi)}{1-\beta}$; Y = C):

$$\begin{array}{lcl} \widetilde{y}_{t}^{j} - \xi_{t} & = & E_{t}[\widetilde{y}_{t+1}^{j-1} - \xi_{t+1}] - \sigma(\widetilde{i}_{t}^{j} - E_{t}\widetilde{\pi}_{t+1}^{j-1}), \ 1 \leq j \leq k \\ \widetilde{y}_{t}^{0} - \xi_{t} & = & -\sigma\widetilde{i}_{t}^{0}, \ j = 0 \end{array}$$

▶ HHs solve this plan at t by **backward induction from** t + k.

Firms K-Horizon Planning

Firms choose P_t^f of good f to maximize:

$$\mathbb{E}_{t}^{f} \left[\sum_{\tau=t}^{t+k} (\alpha \beta)^{\tau-t} \lambda_{\tau} \wp^{f}(\widetilde{p}_{t}^{f}, A_{\tau}) + (\alpha \beta)^{k+1} \vartheta^{f}(\widetilde{p}_{t+k}^{f}) \right]$$

where
$$\widetilde{p}_t^f = \frac{P_t^f \Pi^{\tau-t}}{P_{\tau}}$$
.

► Optimal plan:

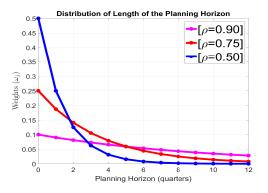
$$\mathbb{E}_{t}^{f} \left[\sum_{\tau=t}^{t+k} (\alpha \beta)^{\tau-t} \lambda_{\tau} \wp_{\widetilde{p}}^{f} (\widetilde{p}_{t}^{f}, A_{\tau}) \frac{P_{t}^{f} \Pi^{\tau-t}}{P_{\tau}} + (\alpha \beta)^{k+1} \vartheta_{\widetilde{p}}^{f} (\widetilde{p}_{t+k}^{f}) \frac{P_{t}^{f} \Pi^{k}}{P_{t+k}} \right] = 0$$

▶ Log-linear approximation (contant $\vartheta^f(\widetilde{p}^f) = \frac{\lambda \wp^f(\widetilde{p}^f)}{1-\alpha\beta}$):

$$\begin{array}{lcl} \widetilde{\pi}_t^j & = & \beta E_t[\widetilde{\pi}_{t+1}^{j-1}] + \kappa(\widetilde{y}_t^j + \xi_t - y_t^*), \ 1 \leq j \leq k \\ \widetilde{\pi}_t^0 & = & \kappa(\widetilde{y}_t^0 + \xi_t - y_t^*), \ j = 0 \end{array}$$

Heterogeneous Planning and Aggregation

- Let ω_j be the fraction of HHs (and Fs) with planning horizon j ($\forall j = 0, 1, 2, ...$). Such that $\sum_j \omega_j = 1$.
- **Exponential Distribution:** $\omega_j = (1 \rho)\rho^j$, $0 < \rho < 1$



Aggregates:

$$\widetilde{\mathbf{y}}_t = (1 - \rho) \sum \rho^j \widetilde{\mathbf{y}}_t^j \qquad \widetilde{\pi}_t = (1 - \rho) \sum \rho^j \widetilde{\pi}_t^j$$

Cyclical Dynamics (Constant Value Functions)

► NK-FH model (cycle):

$$\begin{split} \widetilde{y}_t &= \rho E_t[\widetilde{y}_{t+1}] - \sigma \left[\widetilde{i}_t - \rho E_t(\widetilde{\pi}_{t+1}) \right] \\ \widetilde{\pi}_t &= \beta \rho E_t[\widetilde{\pi}_{t+1}] + \kappa \widetilde{y}_t + \kappa (\xi_t - y_t^*) \\ \widetilde{i}_t &= i_t^* + \phi_{\pi} \widetilde{\pi}_t + \phi_{y} \widetilde{y}_t \end{split}$$

and the baseline NK model if $\rho \to 1$.

In a more compact form:

$$\widetilde{\mathbf{x}}_t = \mathbf{\rho} \mathbf{M} \ E_t \{ \widetilde{\mathbf{x}}_{t+1} \} + \mathbf{N} \mathbf{u}_t$$

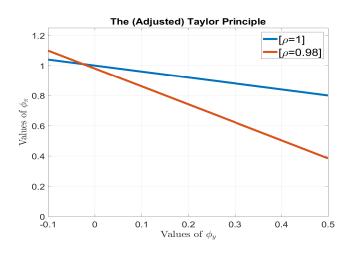
where $\widetilde{x}_t = (\widetilde{y}_t, \widetilde{\pi}_t)'$.

$$M = \frac{1}{\delta} \left(\begin{array}{cc} 1 & \sigma(1 - \beta \phi_{\pi}) \\ \kappa & \kappa \sigma + \beta(1 + \sigma \phi_{y}) \end{array} \right)$$

$$\delta = 1 + \sigma(\phi_{v} + \kappa \phi_{\pi}).$$

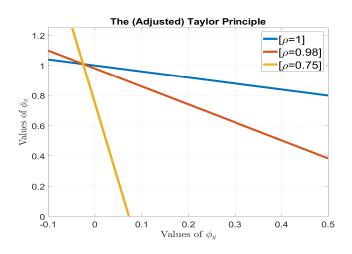
Modified Taylor Principle

$$\frac{1-\rho\beta}{\kappa}\phi_y+\phi_\pi>\rho$$



Modified Taylor Principle

$$\frac{1-\rho\beta}{\kappa}\phi_y+\phi_\pi>\rho$$



Trend Component and Value Function Updating

Updating the value function leads to changes in trends:

$$\begin{split} \widetilde{y}_t &= y_t - \xi_t - \overline{y}_t = (1 - \rho) \sum_j \rho^j (y_t^j - \overline{y}_t^j) - \xi_t \\ \widetilde{\pi}_t &= \pi_t - \overline{\pi}_t = (1 - \rho) \sum_j \rho^j (\pi_t^j - \overline{\pi}_t^j) \\ \widetilde{i}_t &= i_t - \overline{i}_t = (1 - \rho) \sum_j \rho^j (i_t^j - \overline{i}_t^j) \end{split}$$

▶ Without updating the value functions: $\overline{y}_t = \overline{\pi}_t = \overline{i}_t = 0$.

Trend Component and Value Function Updating

- Time-varying trends arise from adjustment in agents' beliefs about the continuation values of their plans.
- ▶ The zero (last bit of planning) condition depends upon the aggregate value functions of households (ν_t) and firms $(\tilde{\nu}_t)$:

$$\overline{y}_{t+k}^{0} = -\sigma \overline{i}_{t+k}^{0} + \nu_{t}
\overline{\pi}_{t+k}^{0} = \kappa \overline{y}_{t+k}^{0} + (1-\alpha)\beta \widetilde{\nu}_{t}$$

For any planning horizon $j \ge 1$ and any date between t and t+j, updating (shooting backward algorithm):

$$\overline{y}_{t+j}^j = \overline{y}_{t+j}^{j-1} - \sigma[\overline{i}_{t+j}^j - \overline{\pi}_{t+j}^{j-1}] \text{ and } \overline{\pi}_{t+j}^j = \beta \overline{\pi}_{t+j}^{j-1} + \kappa \overline{y}_{t+j}^j$$

▶ How do the value functions evolve? Agent's learning.

Value Function Updating: Learning

Constant-gain learning for HHs and Firms:

$$\begin{array}{rcl} v_{t+1} & = & (1 - \gamma)v_t + \gamma v_t^{\text{est}} \\ \widetilde{v}_{t+1} & = & (1 - \widetilde{\gamma})\widetilde{v}_t + \widetilde{\gamma}\widetilde{v}_t^{\text{est}} \end{array}$$

The parameters γ and $\widetilde{\gamma}$ are the constant (learning) gains.

- ν_t^{est} and $\widetilde{\nu}_t^{est}$ are the estimated value functions from period-t decision.
- ► The **continuation values** depend upon (a coarse description of states) aggregate information acquired at time *t*

$$\begin{array}{lll} \boldsymbol{v}_t^{\text{est}} &=& \boldsymbol{y}_t - \boldsymbol{\xi}_t + \sigma \boldsymbol{\pi}_t \\ \boldsymbol{\widetilde{v}}_t^{\text{est}} &=& (1 - \alpha)^{-1} \boldsymbol{\pi}_t \end{array}$$

Monetary Policy

Systematic response to cyclical components

$$\widetilde{i}_t = i_t^* + \phi_\pi \widetilde{\pi}_t + \phi_y \widetilde{y}_t$$

Response to trends (time-varying intercept):

$$\overline{i}_t = \overline{\phi}_y \overline{y}_t + \overline{\phi}_\pi \overline{\pi}_t$$

with $\overline{\phi}_{_{Y}} \geq$ 0, $\overline{\phi}_{\pi} \geq$ 0.

▶ Testable implications: $\phi_y = \overline{\phi}_y$ and $\phi_\pi = \overline{\phi}_\pi$.

$$i_t = \overline{i}_t + \phi_{\pi} \widetilde{\pi}_t + \phi_{y} \widetilde{y}_t + i_t^*$$
or
 $i_t = \phi_{\pi} \pi_t + \phi_{y} y_t + i_t^*$

Aggregate Equilibrium Dynamics

Aggregate dynamics with learning:

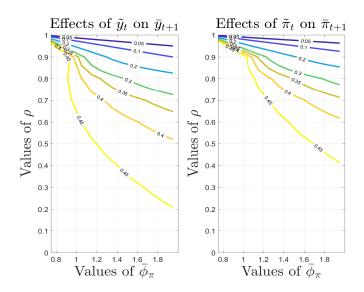
$$\widetilde{x}_{t} = \rho M E_{t} \{\widetilde{x}_{t+1}\} + N u_{t}
\overline{x}_{t} = F \overline{x}_{t-1} + (1 - \rho) \gamma Q \widetilde{x}_{t-1}
\widetilde{x}_{t} = x_{t} - \overline{x}_{t}$$

Assuming $\gamma = \tilde{\gamma}$, then Q becomes:

$$Q = rac{1}{\Delta} \left(egin{array}{cc} 1 - eta
ho & \sigma (1 - eta \overline{m{\phi}}_{m{\pi}}) \ \kappa & \kappa \sigma + (1 -
ho + \sigma \overline{m{\phi}}_{m{y}}) eta \end{array}
ight)$$

with
$$\Delta = (1 - \beta \rho)(1 - \rho + \sigma \overline{\phi}_y) + \kappa \sigma (\overline{\phi}_{\pi} - \rho)$$
.

Monetary Policy and the Passthrough from Cycle to Trend



Estimation Using Aggregate Data

Data and Estimation

Estimate the model over sample 1966:Q1–2007:Q4, with three observables:

Output Growth_t =
$$\mu^Q + y_t - y_{t-1}$$

Inflation_t = $\pi^A + 4\pi_t$
Interest Rate_t = $\pi^A + r^A + 4i_t$

- Period with notable low-frequency variation in these time series.
- We allow for three AR(1) shocks: Technology, Preferences, Monetary Policy.
- We estimate the vector of parameters of the model, θ , using Bayesian techniques.

Key Parameters of the Estimated Models

| Model | Parameters | | |
|--------------------------|---|--|---|
| Туре | Estimated | Fixed | Not Identified |
| Forward | ϕ_{π}, ϕ_{ν} | ho=1 | γ , $\widetilde{\gamma}$, $\overline{\phi}_{\pi}$, $\overline{\phi}_{\nu}$ |
| Stat. Trends | AR(1) trends | ho=1 | γ , $\widetilde{\gamma}$, $\overline{\phi}_{\pi}$, $\overline{\phi}_{y}$ |
| FH Baseline | $\rho, \gamma, \phi_{\pi}, \phi_{y}$ | $\phi=\overline{\phi}$, $\gamma=\widetilde{\gamma}$ | - |
| FH $-\widetilde{\gamma}$ | $ ho$, γ , $\widetilde{\gamma}$, ϕ_{π} , ϕ_{γ} | $\phi=\overline{\phi}$ | - |
| $FH-\overline{\phi}$ | $\rho, \gamma, \phi_{\pi}, \phi_{y}, \overleftarrow{\phi}_{\pi}, \overline{\phi}_{y}$ | $\gamma = \widetilde{\gamma}$ | - |

Key Parameters of the Estimated Models

| Model | Parameters | | |
|--------------------------|--|--|---|
| Туре | Estimated | Fixed | Not Identified |
| Forward | ϕ_{π} , ϕ_{ν} | ho=1 | $\overline{\gamma}$, $\widetilde{\gamma}$, $\overline{\phi}_{\pi}$, $\overline{\phi}_{\gamma}$ |
| Stat. Trends | AR(1) trends | ho=1 | γ , $\widetilde{\gamma}$, $\overline{\phi}_{\pi}$, $\overline{\phi}_{y}$ |
| FH Baseline | $ ho, \gamma, \phi_{\pi}, \phi_{\gamma}$ | $\phi=\overline{\phi}$, $\gamma=\widetilde{\gamma}$ | - |
| FH $-\widetilde{\gamma}$ | $ ho$, γ , $\widetilde{\gamma}$, ϕ_{π} , ϕ_{y} | $\phi=\overline{\phi}$ | - |
| $FH-\overline{\phi}$ | $\rho, \gamma, \phi_{\pi}, \phi_{y}, \overline{\phi}_{\pi}, \overline{\phi}_{y}$ | $\gamma = \widetilde{\gamma}$ | - |

Selected Parameter Estimates: Posterior Distributions

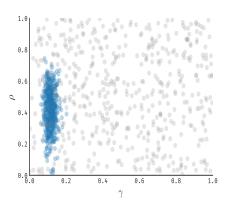
| | Forward | St.Trends | FH Base | FH $-\overline{\phi}$ | FH $-\widetilde{\gamma}$ |
|--|-----------------------|------------------|---------|-----------------------|--------------------------|
| ρ | - | - | | | |
| γ | - | - | | | |
| $\widetilde{\gamma}$ | - | - | | | |
| ϕ_{π} | $\frac{1.54}{(0.24)}$ | 1.49 (0.21) | | | |
| ϕ_y | 0.92 (0.17) | 0.86 (0.19) | | | |
| $\overline{\phi}_{\pi}$ | , , | , , | | | |
| $\overline{\phi}_{\scriptscriptstyle \mathcal{Y}}$ | | | | | |
| log MDD | -758.20 (1.22) | -718.63 (2.16) | | | |

Selected Parameter Estimates: Posterior Distributions

| | Forward | St.Trends | FH Base | $FH	extstyle-\overline{\phi}$ | FH $-\widetilde{\gamma}$ |
|--|--------------------|----------------|-----------------------|-------------------------------|--------------------------|
| ρ | - | - | 0.50 (0.13) | 0.42 (0.13) | 0.46 (0.13) |
| γ | - | - | 0.13 (0.03) | 0.11 (0.02) | 0.09 (0.05) |
| $\widetilde{\gamma}$ | - | - | - | - | 0.17 (0.06) |
| ϕ_π | 1.54 (0.24) | 1.49 (0.21) | $\frac{1.08}{(0.13)}$ | 0.96 (0.15) | 1.10 (0.14) |
| ϕ_y | 0.92 (0.17) | 0.86 (0.19) | $0.78 \\ (0.16)$ | 0.73 (0.15) | 0.77 (0.16) |
| $\overline{m{\phi}}_{\pi}$ | | | - | 2.03 (0.26) | - |
| $\overline{\phi}_{\scriptscriptstyle \mathcal{Y}}$ | - | - | - | 0.06 (0.06) | - |
| log MDD | -758.20 | -718.63 | -727.01 | -716.54 | -728.27 |
| | (1.22) | (2.16) | (0.94) | (1.34) | (1.19) |

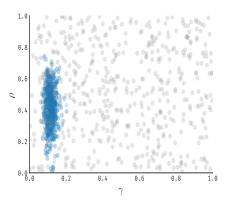
Model Fit

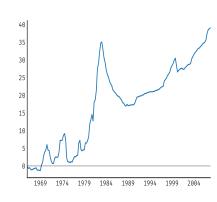
lacktriangle Joint posterior dist. of ho and γ



Model Fit

lacksquare Joint posterior dist. of ho and γ





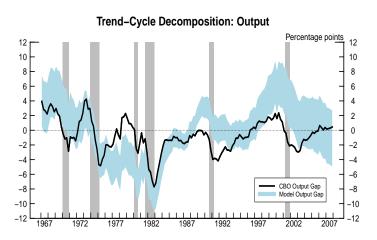
Ranking Overall Fit Alternative Models

$$\widetilde{y}_{t} = \rho E_{t} \{ \widetilde{y}_{t+1} \} - \sigma [\widetilde{i}_{t} - \lambda E_{t} \{ \widetilde{y}_{t+1} \} - r_{t}^{n}]
\widetilde{\pi}_{t} = \beta \rho_{f} E_{t} \{ \widetilde{\pi}_{t+1} \} + \kappa \widetilde{y}_{t} + u_{t}
\widetilde{i}_{t} = \phi_{\pi} \widetilde{\pi}_{t} + \phi_{y} \widetilde{y}_{t} + i_{t}^{*}$$

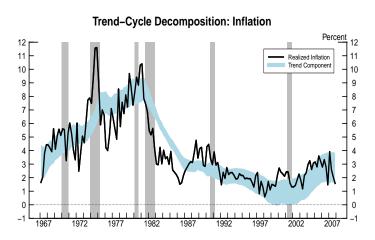
| Model | Log MDD | | |
|--------------------------|---------|------|--|
| Туре | Mean | Std. | |
| $FH-\overline{\phi}$ | -716.54 | 1.34 | |
| Stat. Trends | -718.63 | 2.16 | |
| FH Baseline | -727.01 | 0.94 | |
| FH $-\widetilde{\gamma}$ | -728.27 | 1.19 | |
| Angeletos/Lian/Gabaix | -737.00 | 0.95 | |
| Hybrid NK | -734.24 | 1.55 | |
| Forward | -758.20 | 1.22 | |

Aggregate Implications: Trend vs. Cycle

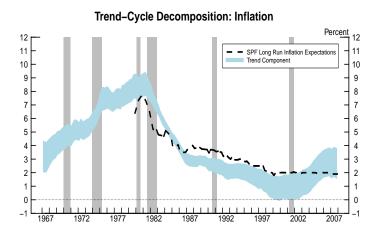
Trend-Cycle Decomposition: Output



Trend-Cycle Decomposition: Inflation



Trend-Cycle Decomposition: Inflation Expectations (SPF)

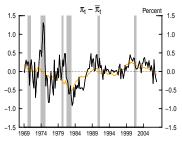


Aggregate Implications: Sources of Business Cycle

Historical Counterfactuals: Monetary Policy

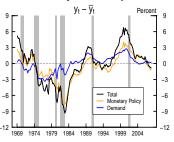
Trend and Cycle of Output and Inflation: Historical Counterfactuals

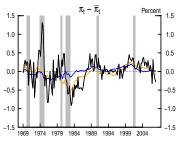




Historical Counterfactuals: Aggregate Demand

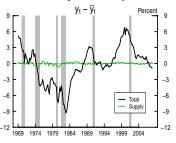
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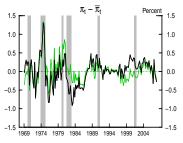




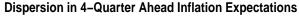
Historical Counterfactuals: Aggregate Demand

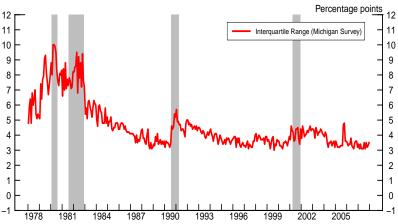
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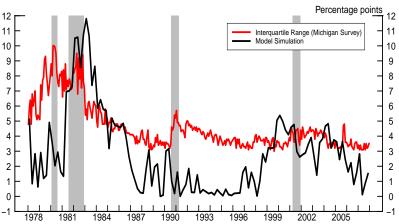


Microeconomic Heterogeneity: Disagreement

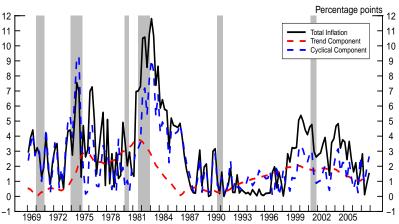




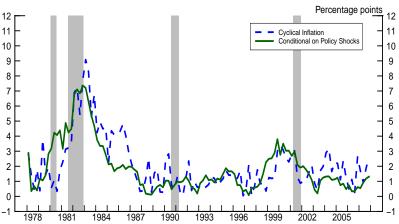






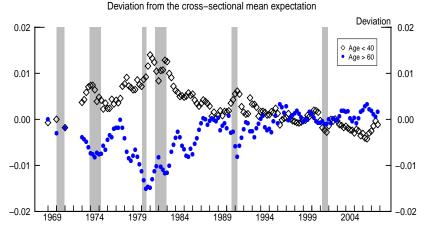


Dispersion in 4-Quarter Ahead Inflation Expectations



Inflation Experiences (Malmendier-Nagel, QJE 2016)

4-Quarter Ahead Inflation Expectations

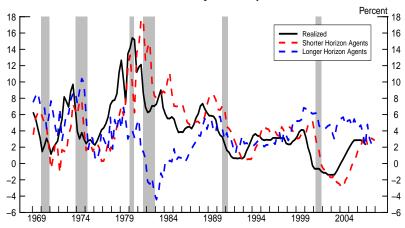


Inflation Expectations across Planning Horizons



Policy Expectations across Planning Horizons





The Role of Short-Planning Heterogeneity

► How important is the (cross sectional) heterogeneity to explain aggregate dynamics? How do finite-horizon planning "representative agent" models fit aggregate data?

| Model | Log MDD |
|----------------------|---------|
| Heterogeneous Agents | -716.5 |
| Rep. Agent: | |
| K = 0 | -720.9 |
| K = 1 | -715.9 |
| K = 2 | -726.2 |
| K = 3 | -734.7 |

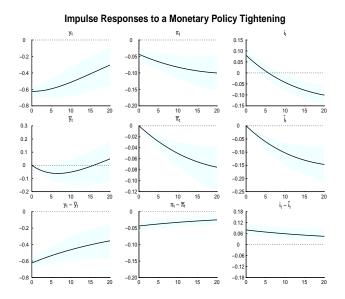
- ▶ Does a flexible "distribution function" help in fitting aggregate dynamics?
 - ▶ Hard to identify *only* with aggregate data.
 - Important to use the cross sectional variation over time on individuals' expectations.

Final Remarks

- ► The FH model outperforms RE versions of the (hybrid) New Keynesian (with intrinsic persistence elements) as well as other behavioral macro models.
- ▶ FH model can be used and extended in several directions:
 - To bring data on individuals' expectations to evaluate the underlying assumptions.
 - To study the heterogeneous implications (on expectations) of alternative MP strategies.
 - ► To explore the effects of FH planning on firms' investment decisions and capital accumulation.

Thank you

Impulse-Responses: Monetary Policy Tightening



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